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Additional Information

1 Are Digital Twins Improving Urban-Water Systems Efficiency and Sustainable Development Goals?

2

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13

14 **Abstract**

15 A digital twin is a tool, which enables a real-time simulation of the water systems and therefore, the water

16 managers can make a decision in the management of the water system over time. The use of these new

17 interaction tool implies the improvement of the awareness of the whole system and it lies in improving the

18 sustainability and efficiency of the water systems with the integration of measurements. The research

19 proposed a methodology to integrate GIS and water models, being the main goal the integration of social,

20 economic, environmental and technical issues. This integration enables improvement in the accuracy and

21 reliability of data and it increases the performance of water systems. This study proposes a pressure-

22 reduction strategy and the implementation of pumps as turbines (PATs), applicable in Sta Cruz, Madeira

23 water system. The use of the developed digital twin model assures a decrease of 3.3 hm³ in water-demand

24 volume, increasing renewable generation by micro-hydropower up to 1.2 GWh. These actions would result

25 in savings above 1.5 M€, decreasing around 530 tons of CO₂ emissions each year. The consideration of

26 these values implies the improvement of different indicators, which allows the evaluation of different

27 targets linked to sustainable development goals (SDGs).

28 **Keywords:** digital twin, leakage control, micro-hydro production, pump as turbines (PATs), urban-water
29 systems, sustainable development goals (SDGs), water systems efficiency

30

31 **1 Introduction**

32 Water losses in water supply and distribution systems, induced mainly by the excess of pressure
33 and a poorly integrated network management, place an immense economic burden on infrastructure grants,
34 as well as the sustainability of water resources, particularly in the context of climate change (Pahl-Wostl
35 2007, Covelli et al. 2016). Increasing the efficiency of water provision by optimizing and rehabilitating
36 existing infrastructures, alternating flow patterns in water networks and control devices, and creating
37 control tools to improve the system behavior and to avoid water and energy wastage, it is an essential
38 measure to be implemented. Therefore, water losses and the associated water–energy nexus should be
39 mitigated to achieve social, economic, and environmental sustainability (Ringler et al. 2013, Bhaduri et al.
40 2015). If effective measures are not implemented, the water supply will be insufficient for essential needs,
41 and energy costs will become prohibitive.

42 A digital-twin hydraulic model can be created, as a crucial and useful management tool that
43 enables the understanding the correlation between different system variables, such as the evolution of
44 pressure patterns, the performance of operating devices, the change of flow velocities and direction, and
45 headlosses occurring in real-time, depending on the operating conditions that change over extended periods.
46 The more information fed into the model, which requires ‘big data’ management, the more results,
47 predictions, and analyses of performance can be implemented.

48 A sustainable development requires new actions to be considered in different social frameworks
49 (Ruggerio 2021) that are important to the water cycle (Yang et al. 2021). The scarcity of water resources
50 and the increasing of economic and environmental costs of non-renewable energies have initiated a search
51 for novel alternatives (Askari Fard et al. 2021). Some of these alternatives focus on using renewable systems
52 such as photovoltaic panels or wind turbines to meet the energy demand.

53 The use of these new tools should be considered since the energy demand will nearly double
54 globally, with water and food demand predicted for 2050 (Karan et al. 2018). The analysis of sustainability
55 could be considered using sustainability indexes, which consider different iteration and relation between

56 water, energy and food nexus (Karan and Asadi, 2018). In this line, the improvement in the data-driven
57 methodologies is able to generate inputs relatively cheaply and easily. The advances in analytics, sensing,
58 transmission, computing and data management are crucial in the upgrading of the urban water systems
59 (Eggimann et al. 2017). Ramos et al. (2020) developed a deep description of different equipments, advices
60 and tools used by water managers in the development of smart water systems, which could help to improve
61 the global efficiency of the water sector. Several authors introduced different techniques to improve the
62 data driven management of water systems, considering different point of views (resources availability,
63 service fails, bad water quality, pressure drops, among others). Xiang et al. (2021) proposed a new adaptive
64 intelligent dynamic water resources planning to improve water efficiency by transforming information into
65 a learner process, improving decision-making based on data-driven by combining numeric AI tools and
66 human intellectual skills. Wu and Liu (2017) developed a deep review of data-driven approaches to improve
67 the knowledge of fails in urban water systems. Wu et al. (2020) presented an adaptive learning rate BP
68 neural network to determine the quality in urban water system, improving the data driven and management.
69 Manny (2022) proposed an analysis data-driven and integrated urban water management to reduce surface
70 water pollution in light of climate change and urbanization impacts in water systems. A spatio-temporal
71 multi-task and multi-view learning was proposed by Liu et al. (2022). All these improvements are linked
72 to applying new technological techniques (Ramos et al. 2020). Then digital decision-making in a water
73 system is based on a digital water twin. It combines knowledge of water infrastructure using information,
74 communication, big data, and social and economic aspects (Pesantez et al. 2022).

75 Using a digital twin as a digital mirror of a real infrastructure allows to improve decision-making.
76 These decisions can be based on two issues: firstly, the analysis of leakages implies to know the all system
77 characteristics and the water system leakages history (Lu et al. 2020). The analysis of leakages has
78 implications for the water-energy nexus, and the control of leakage, which directly reduces the energy
79 consumption in the network, enabling the consciousness of sustainable development goal targets (SDGs)
80 (Ávila et al. 2021); secondly, using digital twins provides a better understanding of the influence of new
81 facility development and enables the recovery of energy or the improvement of the water system
82 management (Bonilla et al. 2022).

83 Water losses in water-distribution systems are directly reflected in economic losses because all
84 associated treatment and transport processes must be conducted. There is a waste of natural resources. Thus,
85 by decreasing the volume of water loss, the pumping energy and the treatment and transport will decrease

86 significantly, improving the whole system efficiency. The analysis of different variables associated with
87 the water and energy consumption, is crucial to define strategies and algorithms to reduce the water and
88 energy use, as well as the integration of renewable systems (Asadi et al. 2020). This complex relationship
89 between them, as well as the social pressure increase, cause the establishment of new approaches to improve
90 the sustainability in water systems (Zare et al. 2020).

91 Energy recovery using micro-hydro solutions can be a valuable strategy for meeting low-cost and
92 long-term–renewable-energy production needs; these solutions include using environmentally sensitive
93 natural or artificial waterfalls integrated into water systems. In developing countries, unconventional
94 solutions are at the forefront of achieving energy self-sufficiency (Ramos and Borga 1999). Reducing water
95 leakage should be considered a new challenge when using water–energy nexus recovery systems (Giustolisi
96 et al. 2008). Water-distribution networks are low-energy efficiency systems because they require high
97 energy levels to satisfy consumption in terms of available pressure, increasing the water leakage volume,
98 the energy consumed by the system, and decreasing the sustainability indices (Morani et al. 2020).

99 Many studies have analyzed the use of micro-hydropower systems in water systems using a pump
100 as a turbine (PAT) (Zhou et al. 2022). The estimation of their main operational curves (Kandi et al. 2021),
101 different regulation strategies (Pugliese and Giugni 2022), and efficiency improvements have been
102 demonstrated in various studies as both advantages and disadvantages (Satish et al. 2021). However,
103 integrating these recovery systems in the digital water twin is not common in published research (Liu et al.
104 2021); this study proposes the development of a procedure in which the digital water model incorporates
105 PAT systems to improve water system management indicators.

106 The excess pressure in water-supply systems (WSS) can be reduced and controlled through
107 specific devices, such as the installation of pressure-reducing valves (PRVs) and flow control valves
108 (FCVs) in strategic locations, and by replacing critical pipe sections to promote good hydraulic system
109 behavior and management (Morani et al. 2020, Ramos et al. 2020). PRVs are typically installed in these
110 systems to control the pressure or the hydraulic head owing to the energy dissipation. These devices can
111 operate in three ways: by closing when the downstream pressure is higher than the value configured in the
112 valve, increasing the headloss until it reaches the established value—if the downstream pressure is lower
113 than the reference value, the valve opens, reducing the headloss—and by closing when the downstream
114 pressure is higher than the upstream pressure, which operates as a check valve by imposing the directional

115 flow change in the water network. Due to the reduced investment made over the years by the concerned
116 water-management entities, water networks operate beyond their useful life and cannot supply the
117 consumption demand resulting from the population growth. This results in supply failures and excessive
118 water losses in systems, which are demonstrated by leakage levels, pipeline ruptures, and reservoir
119 overflows.

120 Water system efficiencies have become a huge concern for management entities, and it should be
121 seen as an opportunity to improve the management of the entire systems (Gleick 2000, Ramos et al. 2021).
122 As a novelty, this study presents a new methodology with practical implementation considerations in line
123 with technical, economic, social, and environmental concerns. The proposed methodology includes the
124 environmental analysis in terms of SDGs targets, as well as the symbiosis between the energy flow available
125 in the water system and micro-hydropower implementation in the improvement of the water-energy nexus,
126 including a feasibility balance with real data applied to a case study. The remaining parts of the paper is
127 organized as follows: Section 2 presents the Methodology, Section 3 states the main results and discusses
128 their relevance in light of similar literature and the economic and environmental benefits of the tested
129 methodology applied in this study. Finally, the main conclusions drawn from this study are summarized in
130 Section 4.

131

132 **2 Methodology**

133 A digital twin uses different technologies to create an interface between a virtual model and a real physical
134 object to send and receive information in real-time (Jiang et al. 2021). The following key technologies must
135 be considered to understand better a digital twin's architecture and infrastructure for a water system:

- 136 - Modelling: Physical and virtual models describe the key features of a water network.
- 137 - Connection: Physical and virtual systems must be constantly connected; moreover, data
138 transmission, conversion, storage, and protection must be performed.
- 139 - Data mining: Data received must be processed, cleaned, and filtered using data analysis
140 techniques and artificial intelligence (AI) to avoid uncertainties or outside correlation values.
- 141 - Interaction and service: Once a simulation is validated, the digital twin must be able to suggest,
142 optimize, and adapt the system processes to induce external changes.

143 Hence, a digital twin is a platform with a set of models that contains different elements depending on the
144 sector or industry. In the water sector, the digital twin should include a water-process model forced to work
145 with suitable boundary conditions, an asset model related to GIS describing physical assets and
146 infrastructure characteristics, topographic representation used to setup and configure the water process, and
147 performance models to generate the metrics required to make decisions, which are usually connected to the
148 enterprise resource planning (ERP) software that enables automated scheduling repairs. The different
149 models must be linked and updated in real-time to represent a complete digital-twin model (Liu et al. 2021).

150 Recently, the Internet of Things (IoT), information and communication technology (ICT) have enabled the
151 facile development of a digital-twin model, which requires the appropriate filtration of information and a
152 big data platform as the input/output of the digital-twin model. The proposed optimization procedure is
153 divided into five stages, each containing different steps. The Methodology is based on routines specifically
154 developed and presented in this study. The procedure is based on the programming in the Epanet toolkit to
155 develop the self-calculation of different iterative procedures. Figure 1 illustrates the steps involved in this
156 Methodology.

157 Step I focuses on data collection in the study area. Obtaining topographic modeling data using GIS tools is
158 essential; these data include flow-recorded data, volume data, and consumption patterns. With all data
159 acquired, the model is developed in Step II by Epanet (Bonilla et al. 2022). The model must be provided
160 alongside the network physical data and the characteristics of the building (i.e., top elevation, number of
161 floors, and eventual pump existence) to understand the pressure and the water-supply management. Once
162 the model is completed, a calibration procedure is developed (Step III). The technique includes an iterative
163 procedure in which the emitter values are changed to reach the water balance according to the billed and
164 unbilled water and uncontrolled volume of water in the water network (Step IIIa.)

165 Leakages are included to simulate both the real and apparent losses. The leakages enable the discretization
166 of the ratio, both real and apparent, using the following equations:

167
$$\eta_{AL} = \frac{V_{AL}}{V_L}, \quad (1)$$

168
$$\eta_{RL} = \frac{V_{RL}}{V_L}, \quad (2)$$

169 where η_{AL} is the ratio between the apparent and total leakages, V_{AL} is the total volume of the apparent losses
170 in m^3 , η_{RL} is the ratio between the real and total leakages, and V_{RL} is the total volume of the real losses in

171 m³. The apparent losses are uncontrolled leakages in water systems that cannot be measured (Ahmadzadeh
172 et al. 2022).

173 Both consumption flow and apparent losses are included in the nodes of the model (Step IIIb). The values
174 of the consumption nodes are considered as the distribution of the population. The population is considered
175 by Thiessen polygons, which overlap with the BGRI polygons and enable the population value to be
176 assigned to each consumption point. The discharged flow is defined as follows (Hamlehdar et al. 2022):

$$q_j = K_f p_j^\gamma, \quad (3)$$

$$K_f = c \times \sum_{j=1}^M 0.5 \times L_{ij}, \quad (4)$$

177 where q_j is the flow discharged from each node in L/s; K_f is the discharge coefficient; p is the pressure in
178 m w.c.; γ is the pressure exponent that depends on the type of material used in the network (Ávila et al.
179 2022); c is the discharge coefficient; L_{ij} is the length of the pipe between the junctions in meters; and i, j ,
180 and M are the nodes and number of pipes connected to node j .

181 Equations (1) and (2) enable the distribution of real losses at different consumption nodes. It is possible to
182 use an iterative procedure (Step IIIb), which minimizes the simulated values (Step IIIc) compared to the
183 measured values, using a programming tool from the Epanet toolkit. The initial value of the discharge
184 coefficient (K_f) is 10^{-5} according to Adedeji et al. (2017), Ávila et al. (2022). The consumption demand and
185 apparent losses are defined as uniform values. In contrast, the real losses are readjusted by the emitter
186 coefficients. The iterative procedure is completed when the difference between the measured and simulated
187 losses is less than the minimum defined value.

188 When the model is calibrated by completing Step III, the procedure continues until Step IV. This step is the
189 development of the digital-twin operation, in which the model is connected to the rest of the variables that
190 integrate the digital twin. Step IV includes the compatibility analysis of the model according to real data
191 (Step IVa) and the development of the modification of incompatibilities (Step IVb), should they arise in the
192 digital twin. When the model is accurate, the use of the proposal and water balance enables the development
193 of an analysis to characterize sustainable improvements (Step V).

194 Different analyses have focused on energy, economic, and environmental issues. Energy balance considers
195 the water balance and the possibility of using micro-hydropower systems. Knowledge of water balance
196 considers the billed and unbilled values of the water company. This proposal includes active leakage control

197 (ALC). It focuses on locating and repairing the broken points of a water system (i.e., pipes and nodes). It is
198 a preventive action to avoid leakage; therefore, the water manager reduces the loss volume in the water-
199 distribution system. The ALC is developed using district-metered areas (DMA). This enables the
200 identification of areas with a higher volume of water loss and the prioritization of pipes for repair. The
201 meshed system enables easy monitoring and control, and its development is based on Galdiero et al. (2015).
202 Energy improvement is based on the location of the recovery system in the water system. As micro-
203 hydropower reduces pressure in the system, energy recovery and reduction of leakages also occur.
204 The energy recovery is analyzed using PATs. The model considers energy recovery using the following
205 equation:

$$E = \gamma Q H \eta, \quad (5)$$

206 where E is the recovered energy in kW, γ is the specific weight of the fluid in $\frac{kN}{m^3}$, Q is the flow in m^3/s ,
207 H is the head recovered by the PAT in m, and η is the global efficiency of the machine.

208 The viability analysis includes economic indicators that consider all costs and benefits. The feasibility was
209 analyzed by considering the net present value (NPV), benefit/cost ratio (B/C), internal rate of return (IRR),
210 and payback period (T) (Abdelhady 2021).

211 The NPV creates a balance between benefits and costs. This is defined as follows:

$$NPV = R - C - O - P, \quad (6)$$

212 where R is the revenue in €, C represents the capital costs in €, O represents the operational costs in €, and
213 P is the repositioning cost in €.

214 B/C relates the present value of benefits to total costs using the following equation:

$$B/C = \frac{R-O}{C+P}. \quad (7)$$

215 The IRR is the discount rate that makes the NPV equal to zero. A project with a higher IRR value is better
216 if the analysis shows a high value. Finally, T represents the number of years that enable the recovery of the
217 initial investment (the pay-back period). The different targets used to measure the evolution of improvement
218 in the SDGs are listed in Table 1.

219

220 The proposed management includes environmental improvement, considering the reduction of CO₂ and its
221 relationship with the achievement of SDGs. Different targets are linked to improve and contribute to
222 measuring these targets' evolution over time inside water systems. This table shows forty-one indicators
223 that can define the environmental operation of the water-distribution system over time.

224 **3 Results and Discussion**

225 **3.1 Brief description of the case study**

226 The case study analyzed is Santa Cruz, located in Madeira Island. Large difference topographic elevations
227 throughout the network characterize the region. Figure 2 shows the Santa Cruz morphology based on the
228 altimetry provided by the municipality of Santa Cruz (MSC). In an area of 81 km², the altitude varies from
229 0 m to over 1000 m. According to PORDATA, Santa Cruz is the second municipality with more inhabitants
230 in Madeira Island. Located in the Atlantic Ocean, this island is characterized by its significant variable
231 altimetry, and the consequent high slopes of the water distribution network. Figure 2 a) presents Santa Cruz
232 morphology considering the altimetry provided. Figure 2b represents the water supply system, developing
233 from a set of pipelines, reservoirs and sources or springs. The municipality has residential and touristic
234 occupation being supplied by 42 reservoirs, mainly by gravity supply. The water distribution systems are
235 aged, with 437 km of pipes, mainly HDPE (46%) and PVC (27%), and 91% of pipe's diameter is lower
236 than DN140.

237 **3.2 Water–energy balance**

238 The proposed methodology was applied in this case study. It searches the best solution in the management
239 to reduce the non-consumed water by the population. The establishment of active leakages control by the
240 introduction of DMAs enables the improvement both water and energy balance. Besides, the procedure
241 analyzed the implementation of control setup and micro-hydropower systems to improve the pressure
242 management. The new control devices (flow control valves, pressure sensors and micro-hydro systems)
243 reach the upgrading of the efficiency of the SCS, avoiding high-pressure variations and pipe breaks. This
244 procedure helps managers to make decisions on the prioritization of system control and maintenance and
245 refurbishment investments.

246 Table 2 shows the annual volume and unbilled/acquired ratio. This shows that more than 7 hm³ have not
247 been billed recently and represents 74% of the total acquired volume. The billed volume was uniform

248 throughout the study period. Thus, the apparent losses did not change over time. The unbilled volume (UV)
249 increased; therefore, the total loss also increased. If the unbilled authorized consumption (UAC) of 7
250 Mm³/year is not billed, it represents 74% of the total billings. The billed volume has remained stable over
251 the years, which indicates that the apparent losses should not vary substantially and may be due to errors in
252 the measurement devices. The evolution of the UV indicates that the total losses increase significantly. The
253 UAC component, corresponding to MSC expenses and intentionally unbilled water, may correspond to a
254 significant portion of the UV. In conclusion, the portion with the greatest weighting increases the unbilled
255 water volume. It contributes to real losses, which is unacceptable.

256 Santa Cruz WDS sub-subsystems are considered, creating new sub-subsystems and network sectorization
257 by creating DMAs. Figure 3 shows the pressure values in the average, peak, and static scenarios for
258 current and future proposals.

259 The comparison between the current and proposed scenarios shows a reduction of the pressure in the
260 system; therefore, advantages are achieved when the pressure and energy terms are considered. Table 3
261 compares the volumes verified in the existing situation and when the proposed solution is applicable.

262 The implementation of the proposed solutions (Figure 3 and Table 3) allows the reduction of the pressure
263 and UV. These measurements enable a 44% reduction in UV above the new total volume and approximately
264 60% relative to the existing scenario. This enables a reduction of 3,344,679 m³ in the total volume required.
265 The model includes the application of ALC measures and controlling all the system volumes. Considering
266 the data in Table 3 and an average annual energy consumption of 0.36 kWh/m³ of water entering the system,
267 an estimated energy saving of 1,204,084 kWh per year can be achieved, resulting in the proposed solution
268 of the MSC WDS.

269 The model was calibrated using the database related to pressure and flow measurements. It enables the
270 definition of an emitter coefficient to simulate the leakages as well as the definition of consumption patterns
271 in the demand nodes. Once the model was verified, the digital twin was used introducing the different
272 assumptions in terms of pressure reduction and micro-hydro solution.

273 The development of digital twins enables the determination of the value of losses and the resulting costs
274 and savings. In further calculations, the average values for buying and selling water can be considered as
275 0.2954 €/m³ and 0.8502 €/m³, respectively, according to the data provided by the MSC. This estimate of
276 economic savings is due to the reduction in water losses. The consumption volumes do not change and the

277 apparent losses represent 20% of the consumption values. In addition, the buying cost to real water-loss
278 volumes and average selling price to apparent losses were applied. The resulting savings for the MSC are
279 listed in Table 4.

280 Therefore, energy savings greater than 1200 MWh are realized when this digital model is applied. It
281 represents an annual value of 118,000 € exclusively from water that would not have to be wasted in the SC
282 system.

283 Table 5 lists the FCVs considered in the proposed solution, which were located upstream of the existing
284 tanks to regulate the entering flows. This change in SC system is designed to control the inlet flows and
285 regulate the pressures in those locations. The following calculation aims to determine the viability of turbine
286 implementation in the two pilot locations, because the cost associated with the additional elements is a
287 fraction of the overall costs. The resulting energy could be sold or used, resulting in additional cost savings.

288 The analysis considers a constant flow over time. This enables the nominal operation of the recovery
289 systems simulated by the model. The case study analyzed is in a QG tank with higher hydraulic power.
290 According to the available head and flow, the chosen PAT is Etanorm 65-250, with its characteristic curve
291 shown in Figure 4a (head curve) and 4b (efficiency curve).

292 In the study of QG behavior through inflows and outflows, an iterative process is performed to confirm the
293 number of daily hours that the PAT would have to work to ensure no overflow or lack of available water
294 volume occurs, considering that the working period of the turbomachine must not experience interruptions.
295 The iterative procedure establishes that the PAT works for approximately 16 h daily, operating an annual
296 volume of 496,437 m³. It is installed with a power of 5.2 kW and annual recovered energy above 31 MWh.
297 The use of these systems implies a reduction of more than 18 tons of CO₂, according to the relationship
298 between generated power and CO₂ emissions (Han et al. 2022).

299

300 **3.3 Viability and environmental analysis**

301 The viability of the recovery systems is analyzed by considering only the recovered energy as a source of
302 income, meaning that no costs or benefits from the water losses and savings previously mentioned will be
303 considered. For this economic analysis, 20 years is considered as the lifespan of the mechanical equipment
304 of the PAT. The installation cost is 2% of the annual maintenance cost (Punys and Jurevičius 2022), and

305 three different discount rates of 6%, 8%, and 10% are used. The energy value is considered 0,098 €/kWh
306 from the ERSE database, as the hydraulic model development is based on the Santa Cruz 2018 water
307 balance values. Table 7 presents the feasibility analysis results.

308 According to the presented results, an average of seven to nine years is required for the payback of the
309 hydropower technology, considering its installation independent of the remaining changes in the Santa Cruz
310 water network. This represents a worst-case scenario because in the present year, from 2022 onwards, due
311 to the increasing energy costs it would represent a higher value, resulting in a higher value of annual revenue
312 and, consequently, a lower payback period.

313 In a last situation, the implementation of the present study synchronized with ALC measures could reach
314 an investment return period of 5.5 years, corresponding to the 2,536,268 €/year saved and to more than 7
315 Mm³ water saving every year. If energy savings are also considered, a value of 2,787,240 € would be saved
316 annually, reducing the payback period to less than five years and avoiding more than 500 tons of CO₂
317 emissions.

318 **4 Conclusions**

319 The development of a digital twin in the MSC system enabled to identify its weaknesses, locate and fix the
320 hydraulic problems, solve the overall excessive pressure scenario, and prevent the lack of supply to some
321 customers and avoid periodic ruptures occurring in the system. This research proposed a practical procedure
322 for water managers to replicate their water systems developing a DT model, enabling the evolution of
323 performance indicators and commensurate targets in the SDGs. This procedure identified forty-one
324 indicators that can be applied.

325 In terms of water balance, after implementing the proposed solution, an annual saving of 3,344,679 m³
326 would be realized, representing 1204 MWh in annual energy savings, representing an estimated economic
327 reserves of more than 1.5 M€ and more than 530 tons/year of CO₂ emissions. Furthermore, if ALC is
328 performed in subsequent years until 15% is achieved—the theoretical optimum proposed level through the
329 PEAASAR II Portuguese target—it would represent an estimated annual saving of more than 2.7 M€ and
330 7 Mm³ of water.

331 In addition, the investigated source of micro-renewable energy upstream of a reservoir bears an
332 insignificant cost when included in the overall network rehabilitation costs, achieving a significant amount
333 of energy recovery and saving of 3050 €/year and 18 tons/year of CO₂ emissions.

334 The research proposes a conventional with AI tools and improvement methods in a digital twin model. The
335 main goal allows the leakage reduction, the system efficiency increases and therefore, the improvement of
336 the water balance and profits of the company. The strength of this research is focused in short-term requiring
337 the development of tools that enable the measurement of different SDG targets through sustainable
338 strategies. These strategies consider technical, social, and economic aspects to promote the incorporation
339 of a weighted multi-criteria model. This DT enables water systems to adapt to climate change and increase
340 their resilience. Under this research lines, the incorporation of different algorithms, to manage the data-
341 driven in the water-energy management allows the evaluation of different SDGs targets over time.

342 **Abbreviations and symbols**

343 AL - Apparent Losses

344 ALC - Active Losses Control

345 AMI - Advanced Metered Infrastructures

346 B - Benefit

347 BGRI - Geographic Referenced Information Database

348 BOD - Biochemical Oxygen Demand

349 BV - Billed values

350 C - Costs

351 CARL - Current Annual Real Losses

352 CMMS - Computerised Maintenance Management System

353 DMA - District Metered Area

354 DT - Digital Twin

355 ERSAR - Water and Wastewater Regulatory Entity

356 ERSE - Energetic Services Regulatory Entity

357 FCV - Flow Control Valve

358 GIS - Geographic Information Systems

359 IoT - Internet of things

360 ICT - Information and Communications Technologies

361 ILI - Infrastructure Leakage Index

362 IRR - Internal Rate of Return

363 MSC - Municipality of Santa Cruz
364 NPV - Net Present Value
365 PAT - Pump as Turbine
366 PEAASAR II- Portuguese Supply and Residual Water Strategic Plan
367 PI - Performance Indicators
368 PRV - Pressure Reducing Valve
369 RL - Real Losses
370 SCS - Santa Cruz Systems
371 SWG - Smart Water Grid
372 T - Payback Period
373 TV - Total values
374 UAC - Unbilled Authorized Consumption
375 UARL - Unavoidable Annual Real Losses
376 UV - Unbilled Values
377 WB - Water Balance
378 WDS - Water-Distribution System
379 WWTP – Wastewater Treatment Plant

380

381 **Author contributions:** Conceptualization, Methodology, and software: HMR and A.K.; validation and
382 formal analysis: HMR, A.K., and MPS; writing—original draft preparation, writing—review and editing,
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